

Regresión no paramétrica para datos funcionales

Liliana Forzani ¹ Ricardo Fraiman ² Pamela Llop ³

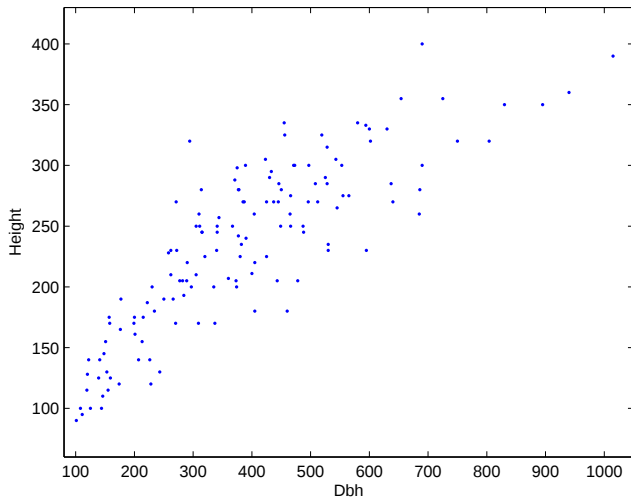
¹Instituto de Matemática Aplicada del Litoral (IMAL) - CONICET - Universidad Nacional del Litoral.

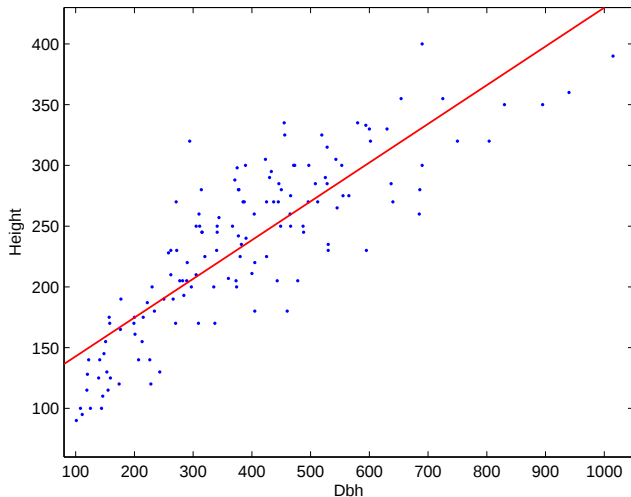
²Departamento de Matemática y Ciencias, Universidad de San Andrés, Argentina
- CMAT, Universidad de la República, Uruguay.

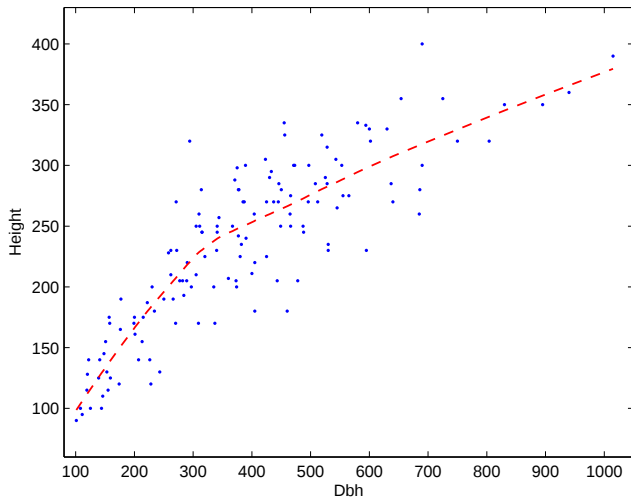
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IMAL 2010

The data file `ufcgf.txt` gives the diameter `Dbh` in millimeters at 137 cm perpendicular to the bole, and the Height of the tree in decimeters for a sample of grand fir trees at Upper Flat Creek, Idaho, in 1991, courtesy of Andrew Robinson.







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Regresión en $\mathbb{R}^d$

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Estimar el operador de regresión $\eta : \mathbb{R}^d \rightarrow \mathbb{R}$ en el modelo

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$$(X_i, Y_i)_{i=1}^n \in \mathbb{R}^d \times \mathbb{R}, \text{ i.i.d. } \longleftrightarrow (X, Y)$$

$$\hat{\eta}(X) \doteq \sum_{i=1}^n W_{ni}(X) Y_i.$$

Teorema (Stone 1977)

(i) Existe $c > 0$ tal que, para toda f , $\mathbb{E}(f(X)) < \infty$,

$$\mathbb{E} \left(\sum_{i=1}^n W_{ni}(X) f(X_i) \right) \leq c \mathbb{E}(f(X)),$$

(ii)

$$\lim_{n \rightarrow \infty} \mathbb{E} \left(\sum_{i=1}^n W_{ni}(X) \mathbb{I}_{\{\|X_i - X\| > a\}} \right) = 0 \quad \forall a > 0,$$

(iii)

$$\lim_{n \rightarrow \infty} \mathbb{E} \left(\max_{1 \leq i \leq n} W_{ni}(X) \right) = 0.$$

Entonces,

$$\lim_{n \rightarrow \infty} \mathbb{E} ((\eta(X) - \hat{\eta}(X))^2) = 0.$$

Ejemplo: Estimador de k -vecinos más cercanos.

$$\hat{\eta}(X) = \sum_{i=1}^n W_{ni}(X) Y_i,$$

$$W_{ni}(X) = \begin{cases} 1/k & \text{si } \|X_i - X\| \leq \|X_{(k)} - X\|, \\ 0 & \text{en otro caso.} \end{cases}$$

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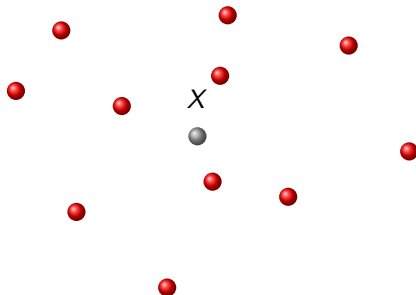
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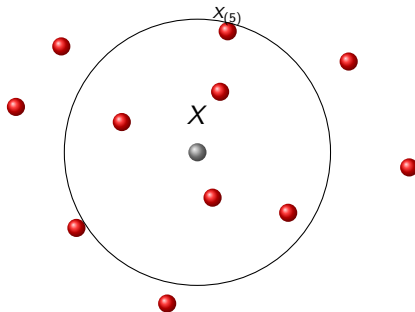


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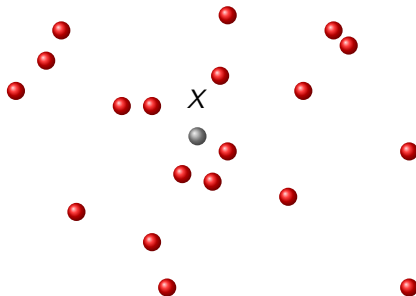
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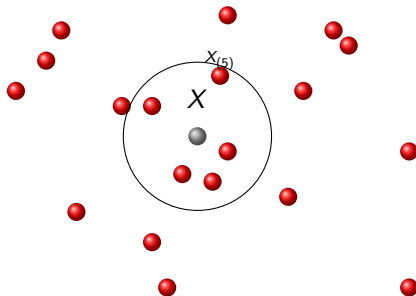


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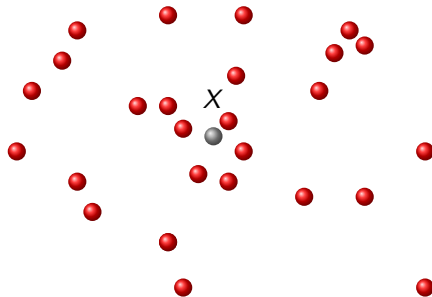
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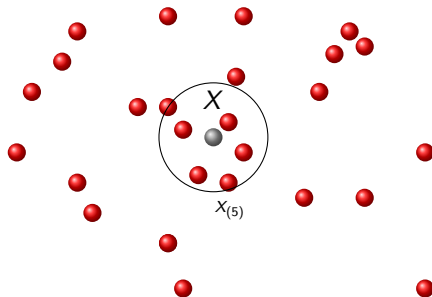
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(iii) η puede aproximarse en L^2 por funciones uniformemente continuas y acotadas.

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Lema

Si $k \rightarrow \infty$ and $k/n \rightarrow 0$, entonces

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Simulación

$$Y = \eta(\mathcal{X}) + e,$$

- $e \sim \mathcal{N}(0, \sigma)$
- $\eta(\mathcal{X}) = \int_0^1 \mathcal{X}(t) dt,$
- $\mathcal{X}(t) = W(t)$
 - $W(0) = 0,$
 - $W(t) - W(s) \sim \mathcal{N}(0, t - s)$ para $0 \leq s \leq t.$

Estimador basado en la medida de ocupación

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Estimador basado en la medida de ocupación

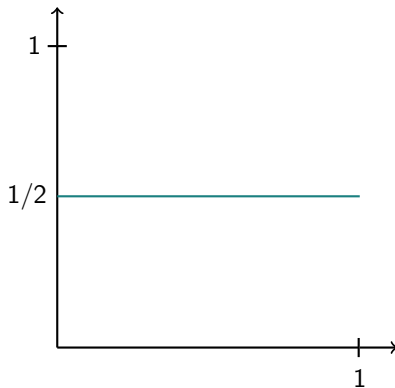
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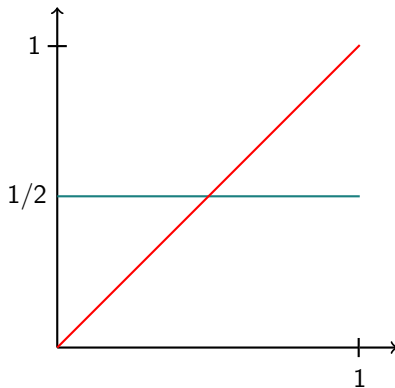
$$\hat{\eta}(\mathcal{X}) \doteq \sum_{i=1}^n W_{ni}(\mathcal{X}) Y_i,$$

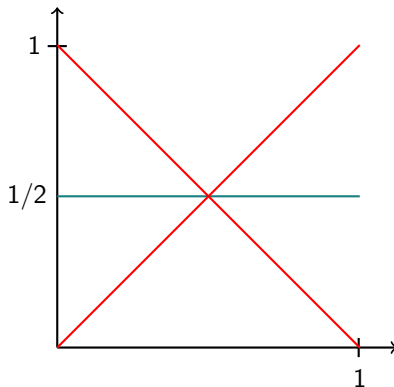
$$W_{ni}(\mathcal{X}) = W_{ni}(\mathcal{X}, \mathcal{X}_1, \dots, \mathcal{X}_n) = \frac{1}{k_n} \int_T \mathbb{I}_{\{t: |\mathcal{X}(t) - \mathcal{X}_i(t)| \leq H_{n,\mathcal{X}}\}} dt.$$

$$k_n \rightsquigarrow H_{n,\mathcal{X}}, \quad k_n = \sum_{i=1}^n \int_T \mathbb{I}_{\{t: |\mathcal{X}(t) - \mathcal{X}_i(t)| \leq H_{n,\mathcal{X}}\}} dt.$$

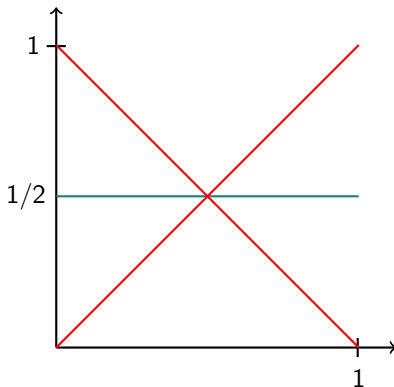


Un nuevo método

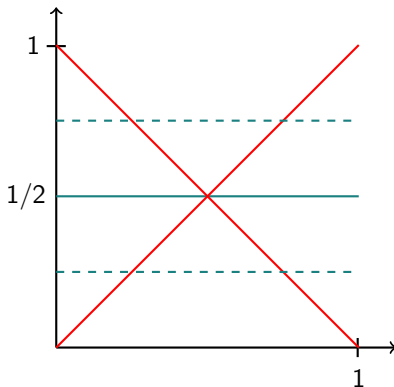




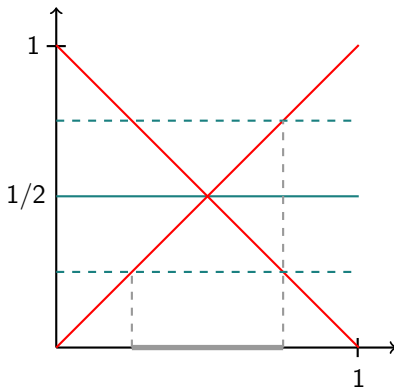
$$\text{Si } k_2 = 1, \text{ ¿} H_{2,\mathcal{X}}, \sum_{i=1}^2 \int_T \mathbb{I}_{\{t: |\mathcal{X}(t) - \mathcal{X}_i(t)| \leq H_{2,\mathcal{X}}\}} dt = 1?$$



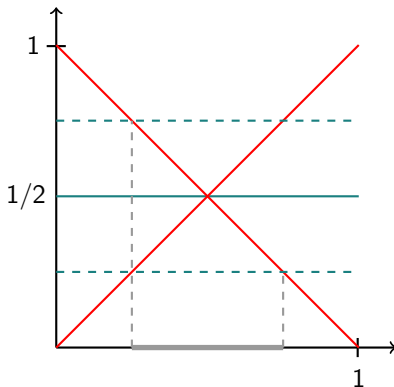
Si $k_2 = 1$, ¿ $H_{2,\mathcal{X}}, \sum_{i=1}^2 \int_T \mathbb{I}_{\{t: |\mathcal{X}(t) - \mathcal{X}_i(t)| \leq H_{2,\mathcal{X}}\}} dt = 1$?



$$\text{Si } k_2 = 1, \text{ ¿} H_{2,\mathcal{X}}, \sum_{i=1}^2 \int_T \mathbb{I}_{\{t: |\mathcal{X}(t) - \mathcal{X}_i(t)| \leq H_{2,\mathcal{X}}\}} dt = 1?$$

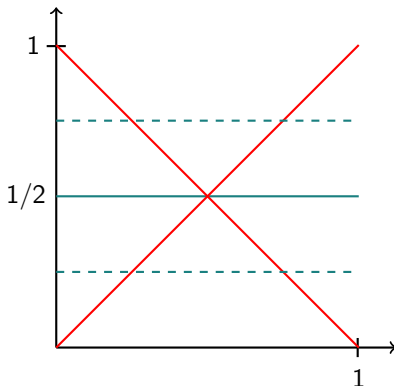

 $\frac{1}{2}$

$$\text{Si } k_2 = 1, \text{ ¿} H_{2,\mathcal{X}}, \sum_{i=1}^2 \int_T \mathbb{I}_{\{t: |\mathcal{X}(t) - \mathcal{X}_i(t)| \leq H_{2,\mathcal{X}}\}} dt = 1?$$



$$\frac{1}{2} + \frac{1}{2} = 1$$

Si $k_2 = 1$, ¿ $H_{2,\mathcal{X}}$, $\sum_{i=1}^2 \int_T \mathbb{I}_{\{t: |\mathcal{X}(t) - \mathcal{X}_i(t)| \leq H_{2,\mathcal{X}}\}} dt = 1$?



Respuesta: $H_{2,\mathcal{X}} = \frac{1}{4}$.

$$\lim_{n \rightarrow \infty} \mathbb{E} ((\hat{\eta}(\mathcal{X}) - \eta(\mathcal{X}))^2) = 0?$$

Simulación

$$Y = \eta(\mathcal{X}) + e,$$

- $e \sim \mathcal{N}(0, \sigma)$
- $\eta(\mathcal{X}) = \int_0^1 \mathcal{X}(t) dt,$
- $\mathcal{X}(t) = W(t)$
 - $W(0) = 0,$
 - $W(t) - W(s) \sim \mathcal{N}(0, t - s)$ para $0 \leq s \leq t.$

Gracias!